

# Space Human Factors and Habitability MIDAS-FAST: Development and Validation of a Tool to Support Function Allocation

Completed Technology Project (2009 - 2013)



## Project Introduction

In this project, the research team 1) developed and validated a model- and simulation-based tool to allow researchers to evaluate various function allocation strategies in space robotics missions and 2) conducted empirical research to investigate human-automation interaction (HAI). The purpose of this tool is to allow human performance researchers and system designers to evaluate potential HAI systems early in the design process. The tool leverages the Man-Machine Integration Design and Analysis System (MIDAS, developed for NASA Ames), and the Basic Operational Robotics Instructional System (BORIS, a NASA Johnson Space Center (JSC) training simulation) to provide MIDAS-FAST (Function Allocation Simulation Tool).

The research proceeded along five partially parallel tracks: (1) developing the function allocation tradeoff model, (2) carrying out empirical human in the loop (HITL) research, (3) developing and (4) validating a computational model of the robotics operator, and (5) implementing the model in the context of the MIDAS-FAST tool. These five major components will be described separately:

1. Function allocation model. A key aspect of function allocation between human and automation is the degree of automation: that is, the relative amount of perceptual, cognitive, and motor "work" carried out by the automation versus human in their collaborative effort in completing task goals. A taxonomy of stages and levels of automation developed by Parasuraman, Sheridan, and Wickens (2000, 2008) describes this degree of automation. One of the important components of the degree of automation is the stage of task information processing at which automation operates to support or replace human activity. Earlier stages involve information acquisition and integration to support situation assessment. Later stages involve action selection and implementation to support task completion. The function allocation tradeoff model underlying FAST proposes that later stages of automation better support routine human-system performance and lower human workload. However these later stages become more problematic if automation fails to perform its functions appropriately, a failure caused in part by the loss of situation awareness. Our review of the literature on human-automation interaction, incorporated into a meta-analysis, supported these tradeoff relationships with statistically significant trends; and the guidance from this FAST tradeoff model have been incorporated into the MIDAS FAST tool (see Items 4 and 5 below).

Using this model, we identified several different types of automation to include in the robotic simulation. These required modifications to the existing BORIS software. Trajectory control automation was implemented in one of three degrees: manual, guided, and automated. To help ensure consistency in experimental participant behavior, we developed 3-segment trajectories that crossed a table (an obstruction) in the operating environment. The first and third segments required movement in 1 axis only; the second segment required movement along 2 axes. Manual trajectories were performed without



Space Human Factors and Habitability MIDAS-FAST: Development and Validation of a Tool to Support Function Allocation

## Table of Contents

Project Introduction	1
Organizational Responsibility	2
Project Management	2
Anticipated Benefits	3
Technology Maturity (TRL)	3
Technology Areas	3
Target Destinations	3
Primary U.S. Work Locations and Key Partners	4
Project Transitions	5
Stories	7
Project Website:	7

# Space Human Factors and Habitability MIDAS-FAST: Development and Validation of a Tool to Support Function Allocation

Completed Technology Project (2009 - 2013)



guidance being given to participants. They were informed of the trajectory to follow, but they were required to determine how to implement it, and to move the arm using the hand controllers. In the automation guided condition, participants were shown a trajectory (or "flight path") to follow. In the autocontrol condition, the trajectory was shown, and the automation executed the trajectory. Hazard alerting and hazard avoidance automation were identified and included. Hazard alerting included color coding to indicate to participants when they had encountered a no-fly zone; hazard avoidance included the color coding as well as stopping the arm to prevent a collision. Camera recommendation logic was also developed. Manual camera control required the operator to make decisions about camera selection, whereas the camera recommendation automation provided a visual alert to suggest a camera switch when needed, and recommended which camera to use. These different types of automation allowed us to research different stages and levels of automation as identified by our framework.

2. Three empirical studies were performed to investigate human performance with different types of robotic arm system automation. The first experiment examined different interface designs, including enhanced (over the current BORIS simulation) graphics for presenting hazards, integrated graphical hazard alerting, and tactile alerting. The second experiment - used for model parameterization and validation - matched the modeling conditions, and examined human performance in conditions with different degrees of automation and with unreliable automation. The third experiment (also used for model validation efforts) investigated adaptive versus adaptable versus fixed automation.

3. The team developed human performance models of scenarios of interest, based on robotic arm task analyses performed in cooperation with subject matter experts (SMEs). The team verified the task analyses by talk-through sessions with SMEs. Human performance model and human-automation interaction predictions were validated in empirical, Human in the loop (HITL) studies identified in Item 2. Results of the validations were used to refine the models. The models included sub models (also referred to as modules) to predict operator visual scanning, operator performance decrements due to poor camera views, and operator decision making. The scanning model is based on SEEV (Saliency, Expectancy, Effort and Value) and the performance impacts of camera view quality were predicted using FORT (Frame of Reference Transformation). SEEV and FORT are relatively mature models, having been developed, refined, and validated under previous NASA research efforts. The decision model was developed specifically for tasks associated with the robotic arm, based on the Generic Robotics Training.

4. A primary goal of this research was to verify and validate our model of the robotic arm operator, to be employed in the function allocation tool, and to collect data that would further validate the Function Allocation Support Tool

## Organizational Responsibility

### Responsible Mission Directorate:

Space Operations Mission Directorate (SOMD)

### Lead Center / Facility:

Johnson Space Center (JSC)

### Responsible Program:

Human Spaceflight Capabilities

## Project Management

### Program Director:

David K Baumann

### Project Manager:

Douglas T Wong

### Principal Investigator:

Angelia Sebok

### Co-Investigators:

Brian F Gore

Nadine Sarter

# Space Human Factors and Habitability MIDAS-FAST: Development and Validation of a Tool to Support Function Allocation

Completed Technology Project (2009 - 2013)



tradeoff model. To accomplish these purposes, data from the Human in the loop (HITL) Experiments 2 and 3 were analyzed, and both models developed and refined.

5. One particular focus of the project was on developing the MIDAS-FAST tool, a prototype model- and simulation-based product that is both usable and useful for researchers, allowing them to easily modify robotic arm scenarios and evaluate different potential automation conditions. This tool offers data entry screens that guide the user through the process of building a scenario. It allows the researchers to specify numerous relevant factors, e.g., operators, tasks, environmental conditions, and function allocation strategy. It offers a visualization capability that provides an animation of the scenario, showing operators interacting with the simulation. The output of the model run includes, in addition to the animation, data files with parameters of interest such as predicted operator situation awareness, workload, visual scanning, camera selection, and trajectory control.

In summary, the MIDAS-FAST project provided a validated model- and simulation based tool for predicting operator performance when working with a robotic arm in different function allocation situations. The function allocation model developed, and the empirical research conducted in this effort were used to identify conditions and provide data development of the tool.

## Anticipated Benefits

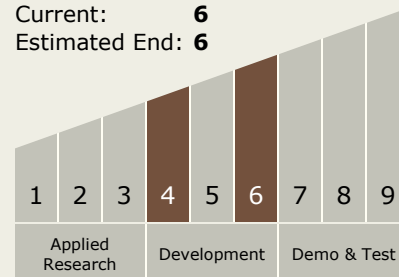
The research provided (and empirically validated) a tool, MIDAS-FAST, to evaluate the effects of human-automation function allocation strategies on human-system performance in robotic tasks involving remote control of a mechanical arm. While the tool was developed specifically for space robotic tasks, we anticipate that the model predictions will also apply to Earth-based robotic tasks.

MIDAS-FAST allows analysts (e.g., researchers, system developers, and concept developers) to enter data regarding the proposed robotic system, allocation of tasks, and the type of automation that is included. The tool uses a variety of sub-models, called modules, to evaluate particular aspects of operator performance (e.g., focus of visual attention, situation awareness, disorientation, and performance decrements due to control-response incompatibilities). The tool provides feedback on predicted operator performance (e.g., time to complete task; trajectory deviations), workload, situation awareness, visual scanning, and camera selection. This will help analysts evaluate and compare potential robotic systems in terms of their predicted effects on operator performance. Model predictions were evaluated and refined with data collected during two human in the loop studies.

Three human in the loop experimental studies, and one meta-analytic literature review conducted during this effort provide empirical data to extend

## Technology Maturity (TRL)

Start: **4**  
Current: **6**  
Estimated End: **6**



## Technology Areas

### Primary:

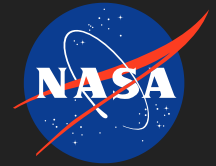
- TX11 Software, Modeling, Simulation, and Information Processing
  - └ TX11.2 Modeling
    - └ TX11.2.3 Human-System Performance Modeling

## Target Destinations

The Moon, Mars

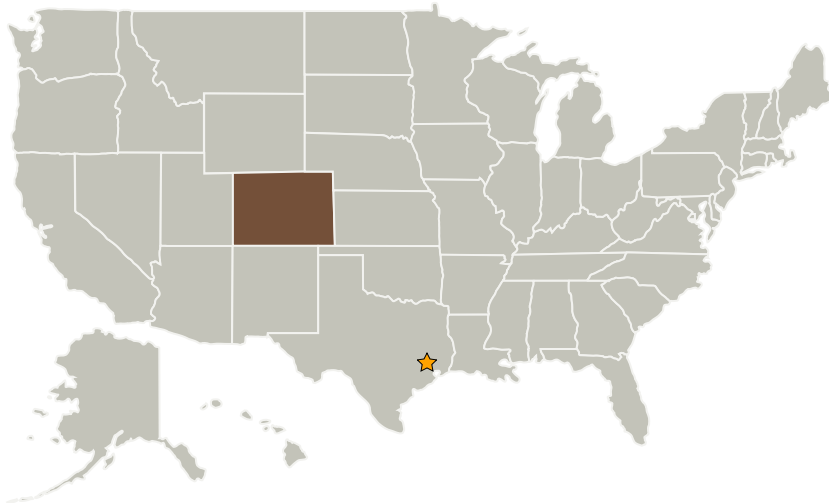
# Space Human Factors and Habitability MIDAS-FAST: Development and Validation of a Tool to Support Function Allocation

Completed Technology Project (2009 - 2013)



the scientific research in human-automation interaction. All four studies have been submitted for publication in either the Human Factors Journal or presentation at the Human Factors and Ergonomics annual conference.

## Primary U.S. Work Locations and Key Partners



Organizations Performing Work	Role	Type	Location
★ Johnson Space Center(JSC)	Lead Organization	NASA Center	Houston, Texas
Alion Science And Technology Corporation	Supporting Organization	Industry	Boulder, Colorado
San Jose State University Research Foundation	Supporting Organization	Academia	San Jose, California
University of Michigan-Ann Arbor	Supporting Organization	Academia	Ann Arbor, Michigan

### Primary U.S. Work Locations

Colorado

# Space Human Factors and Habitability MIDAS-FAST: Development and Validation of a Tool to Support Function Allocation

Completed Technology Project (2009 - 2013)



## Project Transitions



**September 2009:** Project Start

# Space Human Factors and Habitability MIDAS-FAST: Development and Validation of a Tool to Support Function Allocation

Completed Technology Project (2009 - 2013)



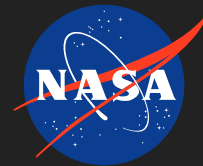
## ✓ April 2013: Closed out

**Closeout Summary:** In the third year of the contract, a number of results were achieved. Briefly summarized, the major accomplishments were the completion of the second and third experimental studies, and model parameterization and validation efforts. This resulted in a number of minor modifications to the human performance model, to simulate more accurately actual operator performance during the robotics tasks. Final development efforts were implemented prior to delivering the software, and a user manual was written. In addition, the project was presented and the MIDAS-FAST tool was shown during a live demonstration as part of an Office of Management and Budget (OMB) deliverable (August 14, 2012). A primary goal of year 3 was to verify and validate our model of the robotic arm operator, to be employed in the function allocation tool, and to collect data that would further validate the Function Allocation Support Tool tradeoff model. To accomplish these purposes, data from the Human in the loop (HITL) Experiments 2 and 3 were analyzed, and both models developed and refined. The nature of Experiment 2, which was the most critical in both validations, will first be described in some detail. HITL simulation: In Experiment 2, 36 participants were first trained extensively to manually operate a simulated version of the robotic arm manipulation task, in which a payload was first raised, then moved across a horizontal "table top", and then lowered to a target destination. While maneuvering, participants needed to avoid proximity with hazards (wall and the table top) and joint states that would freeze the arm motion. In carrying out the task, they were assisted by three forms of automation.

(1) All participants were alerted when proximity to a hazard was violated. (2) All participants were advised, on half the trials, as to the appropriate selection of two of the 4 possible camera views that would provide them with optimal viewing perspective. Finally, (3) participants were divided into three groups receiving different levels of trajectory control automation: none, presence of a 3D guidance path for the correct trajectory (autoguidance), and full autopilot control (autocontrol). In terms of the FAST tradeoff model, the two automation conditions varied in the stage of automation supported (early versus late). Automation of all types functioned correctly through most of the trials. However on the final trial (last few minutes of the 6 hour experiment), both automation systems "failed." Either the line directed the arm too close to a hazard (autoguidance), or the autopilot actually moved the arm into this close proximity with the hazard (autocontrol). In both instances, the collision warning system also failed. Consistent with our FAST tradeoff model, increasing degrees of automation (from none, to autoguidance, to autocontrol) produced progressively better routine performance and lower workload. However, also consistent with the tradeoff model, late stage automation produced significantly worse performance in automation failure management on this final trial, and was associated with a significantly different visual scan path. The camera advisory automation was not failed. The advice of this automation was complied with, and did improve participant's view of the workspace. Validation of robotic arm operator model. We developed a computational model of the robotic arm operator on MIDAS. Because it was MIDAS controlling Boris, we called the model MORIS. MORIS consisted of three primary sub models: (1) A decision model, based on utility theory, chose the best camera views and decided which 4D trajectories to take (XYZ and speed) to reach the ultimate goal and avoid hazards. (2) A spatial cognition model called FORT (Frame of reference transformation) continuously calculated the challenges to spatial cognition caused by different levels of motion ambiguity (portrayed by camera views), control incompatibility (created by misalignment between control motion and perceived display motion), and by visibility challenges within the workplace. (3) a visual scanning model, across the workplace, known as SEEV, which controls simulated eye movement particularly on the basis of the bandwidth of information source changes, their value to the task, and their location in the workplace. MORIS then generated outputs of performance time, trajectory error, workload, camera selection, and scan-based situation awareness. Predictions were different across the three degrees of trajectory automation (none, autoguidance, autocontrol). MORIS model parameters were adjusted so that close fits were obtained between MORIS predictions and the empirical data from the HITL. While the proceeding was essentially "parameterization" of the model, two efforts were made to make true validation: that is, predictions of the model in which the model parameters were not adjusted to maximize the fit with what was predicted. First, we predicted reasonably well, performance of participants in the one condition of experiment 3 that corresponded to one of our conditions in experiment 2. Second, and more significantly, we used MORIS to predict performance of participants in all three conditions in responding to the unexpected automation failures (see above). We did this by modeling, with MORIS, a reduction in the scanning of critical displays, thereby using SEEV to produce an automation complacency prediction. It was this prediction that was validated, with a high correlation between predicted and obtained failure management performance, with the actual performance of participants in the three conditions. In fact, as predicted by the FAST (Function Allocation Simulation Tool) tradeoff model, our complacency measure precisely predicted the poorer performance in the auto-control, compared to the auto-guidance automation. The third experimental study evaluated context-sensitive function allocation. Specifically, four conditions were used to compare operator performance: (1) adaptable automation, where changes in the allocation of functions are initiated by the user, (2) adaptive automation, where the automation triggers changes based on operator or system performance, (3) a hybrid approach, where the system and operator collaborate on selecting and activating automation levels, and (4) a fixed automation approach, in which

# Space Human Factors and Habitability MIDAS-FAST: Development and Validation of a Tool to Support Function Allocation

Completed Technology Project (2009 - 2013)



## Stories

Abstracts for Journals and Proceedings  
(<https://techport.nasa.gov/file/45149>)

Abstracts for Journals and Proceedings  
(<https://techport.nasa.gov/file/45144>)

Abstracts for Journals and Proceedings  
(<https://techport.nasa.gov/file/45145>)

Abstracts for Journals and Proceedings  
(<https://techport.nasa.gov/file/45148>)

Abstracts for Journals and Proceedings  
(<https://techport.nasa.gov/file/45147>)

Abstracts for Journals and Proceedings  
(<https://techport.nasa.gov/file/45146>)

Articles in Peer-reviewed Journals  
(<https://techport.nasa.gov/file/45155>)

Articles in Peer-reviewed Journals  
(<https://techport.nasa.gov/file/45156>)

Articles in Peer-reviewed Journals  
(<https://techport.nasa.gov/file/45151>)

Articles in Peer-reviewed Journals  
(<https://techport.nasa.gov/file/45150>)

Articles in Peer-reviewed Journals  
(<https://techport.nasa.gov/file/45153>)

Articles in Peer-reviewed Journals  
(<https://techport.nasa.gov/file/45154>)

Articles in Peer-reviewed Journals  
(<https://techport.nasa.gov/file/45152>)

Dissertations and Theses  
(<https://techport.nasa.gov/file/45157>)

Papers from Meeting Proceedings  
(<https://techport.nasa.gov/file/45158>)

## Project Website:

<https://taskbook.nasaprs.com>